CSE 332

JUNE 26TH - ANALYSIS OF THE HEAP

• Some problems with EX02 and EX03

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 - Deadline will be extended to Friday

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 - Checkpoint moved to Friday

TODAY'S LECTURE

bigO and analysis

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- bigO and analysis
- Analyzing the heap

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- bigO and analysis
- Analyzing the heap
- Floyd's method

Heap implementation

- Heap implementation
 - Complete

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 - Heap property

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 - Array implementation (0 or 1 indexing)

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 - d-heaps

Algorithm analysis

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 - Want some way for us to compare functions
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 - f(n) = O(g(n)) if there exists some c and n₀ such that f(n) < c*g(n) for some c > 0 and all n > n₀

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- Formally, if f(n) is in O(g(n)) and $\Omega(g(n))$, then f(n) is in $\theta(g(n))$
- Note that the two will have different c and n₀

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 - Sort algorithms into families

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 - Sort algorithms into families
 - O(1): constant
 - O(log n): logarithmic
 - O(n) : linear
 - O(n²): quadratic
 - O(n^k): polynomial
 - O(kⁿ): exponential

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 - Remember that log_a k and log_b k differ by a constant factor?

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 - The constant multiple c lets us organize similar algorithms together.
 - Remember that log_a k and log_b k differ by a constant factor?
 - That makes all logs in the same family

CORRECTNESS ANALYSIS

 How do we show an algorithm is correct?

CORRECTNESS ANALYSIS

- How do we show an algorithm is correct?
 - Need to look at the approach

BINARY SEARCH (AGAIN)

```
public int binarySearch(int[] data, int toFind){
int low = 0; int high = data.length-1;
while(low <= high){</pre>
       int mid = (low+high)/2;
       if(toFind>mid) low = mid+1; continue;
      else if(toFind<mid) high = mid-1; continue;</pre>
      else return mid;
}
return -1;
}
```

BINARY SEARCH CORRECTNESS

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 - Binary search always chooses the correct side
 - End case: low = high



 Let's use these analytical approaches to solve some things about heap functions



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- First, let's do a quick review of heap properties







Is this a heap?



- Is this a heap?
- No. Why

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- Is this a heap?
- Yes, Heap



• Heaps

- Properties
 - Completeness
 - Heap property
- Implementation
 - Array (0 v 1 indexing)

Array property



Array property



Array property



- With 0 indexing:
 - Parent:
 - Left-child:
 - Right-child:

- With 0 indexing:
 - Parent: (i-1)/2
 - Left-child:
 - Right-child:

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- With 1 indexing:
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 - Left-child:
 - Right-child:

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 - Parent: i/2
 - Left-child:
 - Right-child:

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- With 1 indexing:
 - Parent: i/2
 - Left-child: 2i
 - Right-child: 2i+1



• What about for a *d*-heap?

- What about for a *d*-heap?
- Arithmetic changes slightly, but it is still doable



Operations

• Insert: adds a data, priority pair into the heap

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- What are the (worst-case) runtimes for these operations?



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 - What is the height of a heap?

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 - What is the height of a heap? log₂ n

Insert:

- Add the element at the bottom of the tree
- "Percolate up" that element to its correct place
- Adding to the end of a tree? O(1)
- Percolating up? O(height) O(log n)
 - What is the height of a heap? log₂ n



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- Finding in a heap? O(n) Why?
 - Heap property does not give us the divide and conquer benefit
- Percolate up/down? O(log n)
- On average, is it faster to percolate up or down?



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ANALYSIS

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 - Make a new heap and add each element of the array into the heap

ANALYSIS

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- Given an array of length n, how do we make that array into a heap?
- Naïve approach?
 - Make a new heap and add each element of the array into the heap
 - How long to finish?

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What is this summation?


$$\sum_{k=1}^{n} k = \frac{1}{2} n (n+1)$$

FUN FACTS!

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• What does this mean?

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- What does this mean?
- Summing k from 1 to n is still $O(n^2)$
- Similarly, summing log(k) from 1 to n is
 O(n log n)



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• Facts of binary trees



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 Increasing the height by one doubles the number of possible nodes



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Facts of binary trees

- Increasing the height by one doubles the number of possible nodes
- Therefore, a complete binary tree has half of its nodes in the leaves
- A new piece of data is much more likely to have to percolate down to the bottom than be the smallest item in the heap

BUILDHEAP

• So a naïve buildheap takes O(n log n)

BUILDHEAP

• So a naïve buildheap takes O(n log n)

• Why implement at all?

BUILDHEAP

• So a naïve buildheap takes O(n log n)

- Why implement at all?
- If we can get it O(n)!

- Traverse the tree from bottom to top
 - Reverse order in the array

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- Traverse the tree from bottom to top
 - Reverse order in the array
- Start with the last node that has children.
 - How to find? Size / 2
- Percolate down each node as necessary
 - Wait! Percolate down is O(log n)!
 - This is an O(n log n) approach!

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- It is O(n log n), because big O is an upper bound, but there is a tighter analysis possible!
- How far does each node travel (at worst)
 - Leaves don't move at all: Height = 0
 - This is half the nodes in the tree

- It is O(n log n), because big O is an upper bound, but there is a tighter analysis possible!
- How far does each node travel (at worst)
 - 1/2 of the nodes don't move:
 - These are leaves Height = 0
 - 1/4 can move at most one
 - 1/8 can move at most two

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- How far does each node travel (at worst)
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$$\sum_{i=0}^{n} \frac{i}{2^{i+1}} = \frac{2^{-n-1} \left(-n + 2^{n+1} - 2\right)}{2^{n+1} \left(-n + 2^{n+1} - 2\right)}$$

Thanks Wolfram Alpha!

$$\sum_{i=0}^{n} \frac{i}{2^{i+1}} = \frac{2^{-n-1} \left(-n + 2^{n+1} - 2\right)}{2^{n+1} \left(-n + 2^{n+1} - 2\right)}$$

- Thanks Wolfram Alpha!
- Does this look like an easier summation?

$$\sum_{i=0}^{\infty} \frac{1}{2^{i+1}} = 1$$

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This is a must know summation!

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- This is a must know summation!
- 1/2 + 1/4 + 1/8 + ... = 1

$$\sum_{i=0}^{\infty} \frac{1}{2^{i+1}} = 1$$

- This is a must know summation!
- 1/2 + 1/4 + 1/8 + ... = 1
- How do we use this to prove our complicated summation?

 $1/2 + 1/4 + 1/8 \dots + 1/2^n = 1$

- Vertical columns sum to: i/2ⁱ, which is what we want
- What is the right summation?
 - Our original summation plus 1

$$\sum_{i=1}^{\infty} \frac{i}{2^i} = 2$$

$$\sum_{i=1}^{\infty} \frac{i}{2^i} = 2$$

 This means that the number of swaps we perform in Floyd's method is 2 times the size... So Floyd's method is O(n)

NEXT LECTURE

Back to analysis

NEXT LECTURE

- Back to analysis
- Recurrences and analyzing recursive functions